

Simultaneous Confidence Regions for the Multivariate Effective Dose

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Description of Problem

The effective dose (a.k.a. lethal/toxic dose) is the set of covariates which yield a specific response. Let $\mathcal{D} \subset \mathbb{R}^d$ denote the compact domain of the covariates. We assume the a binary response Y is observed and we are interested in sets of the form

$$\text{ED}_{100p}^+ = \{x \in \mathcal{D} : E[Y|X=x] \geq p\} \quad \text{ED}_{100p} = \{x \in \mathcal{D} : E[Y|X=x] = p\}$$

where $x = \{x_1, \dots, x_d\}$ denotes the observed covariates. We assume that the data can be modelled as a logistic regression with

$$\log \left(\frac{E[Y|X=x]}{1 - E[Y|X=x]} \right) = \beta_0 x_0^* + \beta_1 x_1^* + \dots + \beta_k x_k^* = \beta^T x^*,$$

where $x^*(x) : \mathbb{R}^d \mapsto \mathbb{R}^{k+1}$, with $k+1 \geq d$, denotes a smooth function of the covariates. Let $\eta(u) = \log(u/(1-u))$, an increasing function of u . Therefore, we may re-write

$$\text{ED}_{100p}^+ = \{x \in \mathcal{D} : \beta^T x^* \geq \eta(p)\} \quad \text{ED}_{100p} = \{x \in \mathcal{D} : \beta^T x^* = \eta(p)\}.$$

We estimate these using the plug-in estimators

$$\widehat{\text{ED}}_{100p}^+ = \{x \in \mathcal{D} : \widehat{\beta}_n^T x^* \geq \eta(p)\} \quad \widehat{\text{ED}}_{100p} = \{x \in \mathcal{D} : \widehat{\beta}_n^T x^* = \eta(p)\},$$

where $\widehat{\beta}_n$ is the well-known maximum likelihood estimator of β . Our goal is to develop a simultaneous confidence region for ED_{100p}^+ and ED_{100p} .

Calculating the Confidence Regions

Below, we explain how to calculate the $100(1-\alpha)\%$ confidence region for ED_{100p}^+ . A $100(1-\alpha)\%$ confidence region for ED_{100p} is taken as the intersection of the $100(1-\alpha/2)\%$ confidence region for ED_{100p}^+ and the $100(1-\alpha/2)\%$ confidence region for $(\text{ED}_{100p}^+)^c$.

(1) INVERTING SCHEFFE'S BOUNDS

A conservative, and currently only, method developed by Carter et al. (1986) is based on inverting Scheffé's upper bound. Recall that $\lim_n \sqrt{n}(\widehat{\beta}_n - \beta) \sim N_{k+1}(0, \Sigma)$. Then the $100(1-\alpha)\%$ confidence region for ED_p^+ is

$$\text{CR}_{\text{SCH},100(1-\alpha)}^+ = \left\{ x : \widehat{\beta}_n^T x^* \geq \eta(p) - \sqrt{\chi_{\alpha}^2(k+1) x^{*T} \widehat{\Sigma} x^* / n} \right\}.$$

where $\widehat{\Sigma}$ is the usual estimate of Σ . This approach was studied extensively by Li et al. (2008a) in the linear setting.

(2) VOROBÉV QUANTILE

Let \mathbf{A} be a random closed set (RCS) and define $T(K) = P(\mathbf{A} \cap K \neq \emptyset)$. As defined in Molchanov (1990), the p -quantile of \mathbf{A} is $M_p = \cup \{K \in \mathcal{M} : T(K) < p\}$, and we choose $\mathcal{M} = \{\{x\}; x \in \mathbb{R}^d\}$. Then the $100(1-\alpha)\%$ confidence region for ED_p^+ is

$$\text{CR}_{\text{Q},100(1-\alpha)}^+ = M_{\alpha}^c,$$

where M_{α} denotes the quantile of $\widehat{\text{ED}}_p^+$. This is estimated empirically via re-sampling. Heuristically, the confidence region is the collection of points $\{x\}$ such that each falls inside $\widehat{\text{ED}}_p^+$ at least $100(1-\alpha)\%$ of the time.

(3) NEW DEFINITION OF RCS QUANTILE

Let \mathcal{A} be a collection of sets such that $\mathbf{A} \in \mathcal{A}$ almost surely. Fix a set K , and define the random variable $\xi = \rho_H(\mathbf{A}, K)$. Let q_p^{ξ} denote the p -quantile of ξ . The quantile of \mathbf{A} is defined as $M_p^H = \cup \{A \in \mathcal{A} : \rho_H(A, K) \leq q_p^{\xi}\}$, where ρ_H denotes the Hausdorff distance. The $100(1-\alpha)\%$ confidence region for ED_p^+ is then

$$\text{CR}_{100(1-\alpha)}^+ = M_{1-\alpha}^H,$$

where $M_{1-\alpha}^H$ is the quantile of $\widehat{\text{ED}}_p^+$. Draw A_1, \dots, A_n samples of ED_p^+ and calculate $K_n = \cap_{i=1}^n A_i$ with $\xi_{i,n} = \rho_H(A_i, K_n)$. Then the estimate of the confidence region is

$$\cup \{A_i, i = 1, \dots, n : \rho_H(A_i, K_n) \leq \widehat{q}_{n,1-\alpha}^{\xi}\},$$

where $\widehat{q}_{n,p}^{\xi}$ is the quantile estimate from the observations of $\xi_{i,n}$.

Results

We consider the following four models.

model	true value of $\beta^T x^*$	domain
linear	$-6 + 6x_1 + 6x_2$	$[0, 1]^2$
interaction	$-6 + 6x_1 + 6x_2 - 3x_1x_2$	$[0, 1]^2$
quadratic	$-6 + 6x_1 + 6x_2 + 10x_1^2 + 3x_1x_2 + x_2^2$	$[0, 1]^2$
log term	$-10 + 6 \log x_1 + 6x_2$	$[1, 2]^2$

To compare our methods, we ran Monte Carlo simulations to find the empirical coverage probabilities. We simulate ED_{100p} values under the asymptotic distribution of $\widehat{\beta}_n$. The results are shown in the tables below.

p	linear			interaction			quadratic			quadratic (+)			log term		
	(1)	(2)	(3)	(1)	(2)	(3)	(1)	(2)	(3)	(1)	(2)	(3)	(1)	(2)	(3)
0.1	.99	.89	.97	1.00	.89	.97	1.00	.87	.99	1.00	.87	.99	.99	.92	.98
0.5	.98	.86	.94	1.00	.79	.94	1.00	.80	.98	1.00	.81	.97	.99	.87	.96
0.9	.99	.90	.99	1.00	.91	1.00	1.00	.80	.97	1.00	.80	.96	.98	.88	.96

Table 1: Empirical coverage results for 95% confidence regions when simulating from the limiting distribution. The sample size is 360, except in the fourth column (quadratic (+)), where the sample size is 3600. Results not statistically different from 0.95 are shown in bold.

p	linear			interaction			quadratic			quadratic (+)			log term		
	(1)	(2)	(3)	(1)	(2)	(3)	(1)	(2)	(3)	(1)	(2)	(3)	(1)	(2)	(3)
0.1	.18	.12	.17	.26	.16	.21	1.00	.15	.30	1.00	.03	.05	.10	.07	.10
0.5	.19	.13	.20	.27	.17	.27	1.00	.12	.47	1.00	.03	.05	.18	.12	.19
0.9	.18	.12	.19	.12	.09	.12	1.00	.24	.66	1.00	.04	.07	.26	.18	.28

Table 2: Median proportion of the domain covered by the confidence region. The simulations are the same as those shown in Table 1.

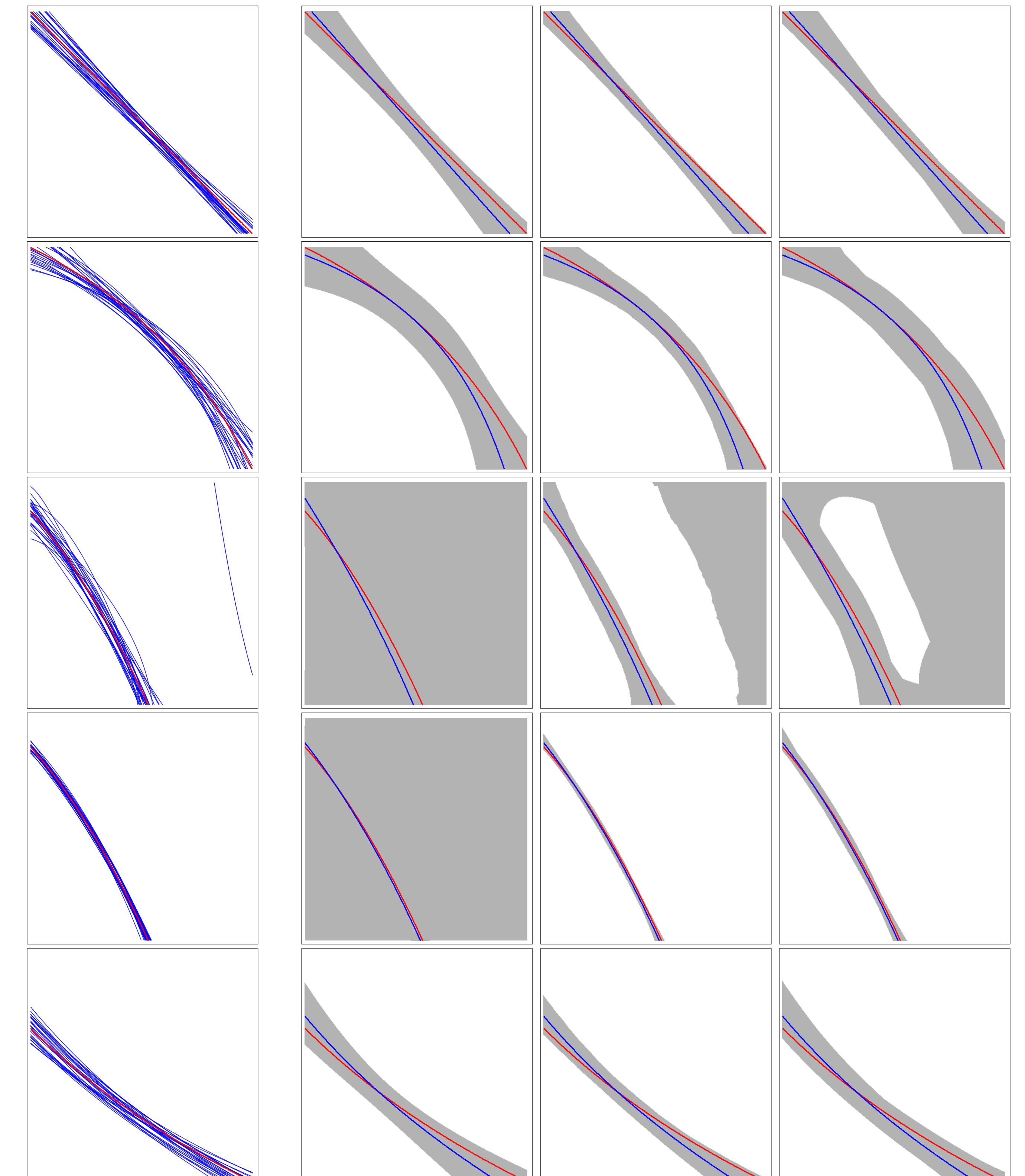


Figure 1: Examples of the different methods: In the leftmost column are sample observations of $\widehat{\text{ED}}_{50}$ and the three columns on the right show the three regions (1) $\text{CR}_{\text{SCH},95}$, (2) $\text{CR}_{\text{Q},95}$, and (3) CR_{95} in light grey, from left to right. Observed values of $\widehat{\text{ED}}_{50}$ are shown in blue, and the true set ED_{50} is shown in red. From top to bottom the models are linear ($n = 360$), interaction ($n = 360$), quadratic ($n = 360$), quadratic ($n = 3600$), and log term ($n = 360$).

References

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